# 12. Incomplete Information

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Complete information assumption implies players know others' payoffs.

#### Examples:

- goal keeper may not really know how effortful it is for the penalty kicker to shoot right instead of left;
- firms may not know other firms' cost structure;
- voters may not know how other voters' preferences;
- consumers may be unsure of how much they value a good;
- investors may not know what is the value of an asset;
- firm may not know how productive a given job candidate is;
- a researcher may not know how difficult a problem they're working on is.

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**Today's agenda:** formalising games of incomplete information and examining applications.

- 1. Motivation
- 2. Bayesian Games
- 3. Bayesian Nash Equilibrium
- 4. Auctions
- 5. Purification Theorem
- 6. Higher-Order Beliefs
- 7. More

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- 2. Bayesian Games
  - Representing Incomplete Information
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#### **Definition**

A game is of **incomplete information** when at least one player does not know the payoff that some player receives from some strategy profile.

How to model uncertainty?

### Harsanyi's modelling insight:

Transform incomplete info game into complete info with Nature moving at start of game.

Realisation of nature's actions determines players' payoffs.

Assumption: CK of prob. distrib. used by Nature.

Players have a belief about others' preferences and there is common knowledge of such beliefs.

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#### **Definition**

A **Bayesian game** is a tuple  $\langle I, A, u, \Theta, \rho \rangle$ , where

- (i) Players: I;
- (ii) Player i's action space  $A_i$ ; Space of action profiles  $a \in A := \times_{i \in I} A_i$ ;
- (iii) Player i's type space  $\Theta_i$ ; Space of type profiles  $\Theta := \times_{i \in I} \Theta_i$ ;
- (iv) Player i's utility/payoff function:  $u_i : A \times \Theta \to \mathbb{R}$ ;  $u := (u_i)_{i \in I}$ ; and
- (v) Probability distribution over players' type profiles:  $\rho \in \Delta(\Theta)$ .

All elements are common knowledge, but each player i only knows their own type  $\theta_i$ , and not the other players' types.

Players privately learn their own type. (WLOG)

#### **Definition**

• Players have **private values** iff  $u_i(a, \theta_i, \theta_{-i}) = u_i(a, \theta_i, \theta'_{-i}) \ \forall \theta_{-i}, \theta'_{-i} \in \Theta_{-i}$ . Otherwise, they have **interdependent values**.

Could also consider alternative notions of incomplete information: e.g., uncertainty over what is the strategy set of the opponent.

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#### **Definition**

A pure **strategy** of player *i* in a Bayesian game is a mapping  $s_i : \Theta_i \to A_i$ .

Strategy specifies action for each possible type.

Player *i*'s expected payoff:  $\tilde{u}_i(s) = \mathbb{E}_{\theta \sim p}[u_i(s_1(\theta_1), s_2(\theta_2), ..., s_l(\theta_l), \theta_i, \theta_{-i})].$ 

Extend  $\tilde{u}_i$  to mixed strategies,  $\sigma_i \in \Sigma_i := \Delta(S_i)$ .

Two classmates, A and B, considering whether to work together.

They work together iff both agree to do so.

If they work alone, payoffs normalised to 0.

If they work together B always gets 10 (improves their grade by 10). However, how much A benefits from working with B depends on B's type.

If B is collaborative (wp  $\alpha$ ), A also gets a payoff of 10. But if B is a shirker (wp  $1 - \alpha$ ), then A gets a payoff of -6.

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	Table: $\theta_B = C$				Table: $\theta_B = S$				
	В				В				
		W	Ν					W	Ν
Α	W	10,10 0,0	0,0			А	W	-6,10 0,0	0,0
	Ν	0.0	0.0				Ν	0.0	0.0

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#### Higher-order uncertainty and belief hierarchy

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Yes, with a **universal type space** (Mertens & Zamir (1985); Bradenburger & Dekel (1993))

Reassuring that Bayesian games are good tool.

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- 3. Bayesian Nash Equilibrium
  - Ex-ante vs Interim perspective
- 4. Auctions
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#### **Definition**

A **Bayesian Nash Equilibrium** of a Bayesian game  $\langle I, A, u, \Theta, \rho \rangle$  is a strategy profile  $s = (s_i)_{i \in I}$  such that  $\forall i, \forall s_i' \in S_i, \tilde{u}_i(s_i, s_{-i}) \geq \tilde{u}_i(s_i', s_{-i})$ .

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NB: consider Bayesian game  $\Gamma$  =  $\langle I, A, u, \Theta, \rho \rangle$  as standard normal-form game  $\tilde{\Gamma}$  =  $\langle I, S, \tilde{u} \rangle$ .

Set of BNE of  $\Gamma$  is the same as set of NE of  $\tilde{\Gamma}$ .

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### Corollary

For any Bayesian game  $\Gamma$  s.t. |I|, |A|,  $|\Theta| < \infty$ ,  $\exists$  Bayesian Nash equilibrium (possibly in mixed strategies).

#### **Ex-ante Perspective:**

- players choose strategies, (distrib. over) mappings from types to actions, to maximise ex-ante expected payoff;
- 2. types are drawn according to  $\rho$ ;
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#### Interim perspective:

- 1. types are drawn according to  $\rho$ ;
- 2. players learn their own types, form beliefs about others' types  $q_i(\cdot \mid \theta_i)$ , and play according to their actions;
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- 4. outcomes and payoffs realise.

Arguably more sensible description of a game of incomplete information.

#### **Definition**

An ex-interim **Bayesian game** is a tuple  $\langle I, A, u, \Theta, q \rangle$ , where

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- (ii) Player i's action space  $A_i$ ; Space of action profiles  $a \in A := \times_{i \in I} A_i$ ;
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- (iv) Player *i*'s utility/payoff function:  $u_i : A \times \Theta \to \mathbb{R}$ ;  $u := (u_i)_{i \in I}$ ; and
- (v) Ex-interim Belief/Prob. distrib. over opponents' type profiles:  $q_i: \Theta_i \to \Delta(\Theta_{-i})$ .

### **Proposition**

A strategy profile  $\sigma$  is a BNE if and only if  $\forall i \in I$  and  $\forall \theta_i \in \Theta_i : \rho(\theta_i) > 0$ ,

$$\mathbb{E}_{\theta_{-i}}[u_i(\sigma_i(\theta_i),\sigma_{-i}(\theta_{-i}),\theta_i,\theta_{-i}) \ | \ \theta_i] \geq \mathbb{E}_{\theta_{-i}}[u_i(\sigma_i'(\theta_i),\sigma_{-i}(\theta_{-i}),\theta_i,\theta_{-i}) \ | \ \theta_i], \ \forall \sigma_i' \in \Delta(A_i)^{\Theta_i}.$$

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NB: Possible to find  $\rho_i \in \Delta(\Theta)$ :  $\rho_i(\theta_{-i} \mid \theta_i) = q_i(\theta_{-i} \mid \theta_i) \quad \forall \theta_i, \theta_{-i}$  (going beyond finite case introduces technical complications).

However: players may not start with **common prior**:  $\rho_i = \rho_j$  for all i, j.

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Common prior necessary for equivalence between ex-ante BNE and interim BNE.

# Ex-Post Bayesian Nash Equilibrium

#### **Definition**

A strategy profile  $\sigma$  is an **ex-post Bayesian Nash equilibrium** iff  $\forall i, u_i(\sigma_i(\theta_i), \sigma_{-i}(\theta_{-i}), \theta) \geq u_i(a_i, \sigma_{-i}(\theta_{-i}), \theta)$ ,  $\forall a_i, \theta$ .

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## Ex-Post Bayesian Nash Equilibrium

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# Strategy-Proofness

Closely related to "**Very weak dominance**":  $s_i : u_i(s_i(\theta_i), a_{-i}, \theta_i, \theta_{-i}) \ge \tilde{u}_i(a_i, s_{-i}, \theta_i, \theta_{-i}), \forall a_i, a_{-i}, \forall \theta$ .

Allows for indifferences.

Also said **Strategy-proofness**, esp. when  $A_i = \Theta_i$ .

You'll hear this term a lot.

## Overview

- 1. Motivation
- 2. Bayesian Games
- 3. Bayesian Nash Equilibrium
- 4. Auctions
  - 2nd-Price Auction
  - Envelope Theorem
  - 1st-Price Auction
  - Revenue Equivalence
- 5. Purification Theorem
- Higher-Order Beliefs
- 7. More

### 2nd-Price Auction

**2nd-Price Auction:** winner pays second highest bid.

$$u_i(a_i, a_{-i}, v_i) = 1\{i \in \arg\max_j a_j\}(v_i - \max_{j \neq i} a_j)/|\arg\max_j a_j|$$

When  $F_i$  is degenerate for every i,  $a_i = v_i$  is weakly dominant for all players (hence a NE?).

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### Independent private values. (What does this mean?)

 $v_i \sim F_i$ ,  $v_i$  independent from other types.

 $s_i$ :  $s_i(v_i) = v_i$  still weakly dominant for all players? Is it a BNE? What does it depend on?

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**Informationally robust** (although perhaps counterintuitive to people.)

**Alternative: Ascending auction.** People understand it better and play weakly dominant strategy more often.

With good reasons: Obviously Strategy-Proof (Li, 2017 AER)

Roughly, worst case scenario better than best-case scenario from deviation.

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Relate effect of parameter on value function to its effect on the objective function.

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Choice set X.

Parameter  $t \in [0, 1]$  (think directional derivative in normed vector space)

Objective function:  $f: X \times [0,1] \to \mathbb{R}$ .

Value function:  $V(t) := \sup_{x \in X} f(x, t)$ ; Maximisers  $X^*(t) := \{x \in X : f(x, t) = V(t)\}$ .

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### Theorem 1 (Milgrom & Segal 2002 Ecta)

Take any  $x^* \in X^*(t)$  and  $t \in [0, 1]$ , and suppose  $f'_t(x^*, t)$  exists.

- (1) For t > 0, if V is left-differentiable at t,  $V'(t^-) \le f'_t(x^*, t)$ .
- (2) For t < 1, if V is right-differentiable at t,  $V'(t^+) \ge f'_t(x^*, t)$ .
- (3) For  $t \in (0, 1)$ , if V is differentiable at t, then  $V'(t) = f'_t(x^*, t)$ .

It would be sufficient to ensure V is differentiable a.e. to get

### Theorem 2 (Milgrom & Segal 2002 Ecta)

Take any  $x^* \in X^*(t)$  and  $t \in [0, 1]$ , and suppose  $f'_t(x^*, t)$  exists.

(1) If  $f(x, \cdot)$  is absolutely continuous for all  $x \in X$  and there is an integrable function  $b: [0,1] \to \mathbb{R}_+$  such that  $|f'_t(x,t)| \le b(t) \ \forall x \in X$  and almost all  $t \in [0,1]$ , then V is absolutely continuous.

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- (2) If, in addition,  $f(x, \cdot)$  is differentiable for all  $x \in X$  and  $X^*$  is nonempty-valued a.e. on [0, 1], then for any selection  $x^*(t) \in X^*(t)$ ,

$$V(t) = V(0) + \int_0^t f_t'(x^*(s), s) ds$$

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Note: V need not be differentiable everywhere (may have kinks).

Back to auctions: 1st-Price Auction: winner pays highest bid.

*I* bidders with valuations  $0 \le v_i$  and  $v_i \sim F$  iid, *F* atomless and absolutely continuous, bounded support  $V_i = [\underline{v}, \overline{v}]$ 

**Bids:**  $a_i \geq 0$ .

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**Bids:**  $a_i \geq 0$ . **Strategies:**  $s_i : V_i \rightarrow \mathbb{R}_+$ 

### **Payoffs**

$$u_i(a_i, a_{-i}, v_i) := \mathbf{1}_{a_i \in \max_{j \in I} \{a_j\}} \frac{1}{|\arg\max_{i \in I} \{a_i\}|} (v_i - a_i)$$

Get zero if do not bid highest.

Get item if bid highest and pay own bid; uniform tie-breaking.

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NB:  $s_i(v_i) = v_i$  is weakly domina**ted** in 1PA!

# Solving for a symmetric PS-BNE $(s^*)_{j \in I}$

Assume: s\* is strictly increasing, differentiable.

 $s^*$  strictly increasing + F atomless  $\implies$  zero prob. of two identical bids.

## Solving for a symmetric PS-BNE $(s^*)_{j \in I}$

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Expected payoff from bidding  $a_i$  given type  $v_i$  and opponents bidding according to  $s^*$ :

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$$u(a_{i}, v_{i}) = F((s^{*})^{-1}(a_{i}))^{|I|-1}(v_{i} - a_{i}).$$

$$\Longrightarrow U(v_{i}) := u(s^{*}(v_{i}), v_{i}) = F((s^{*})^{-1}(s^{*}(v_{i})))^{|I|-1}(v_{i} - s^{*}(v_{i}))$$

$$U(v_{i}) = F(v_{i})^{|I|-1}(v_{i} - s^{*}(v_{i})).$$
(1)

# Solving for a symmetric PS-BNE $(s^*)_{i \in I}$

Assume: s\* is strictly increasing, differentiable. (check later)

 $s^*$  strictly increasing  $\implies s^*(\underline{v})$  wins auction wp0  $\implies U(\underline{v}) = \mathbf{0}$ .

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- (a)  $u(a_i, v_i) = F((s^*)^{-1}(a_i))^{|I|-1}(v_i a_i)$  differentiable in  $v_i$ .
- (b)  $U(v_i) = F(v_i)^{|I|-1}(v_i s^*(v_i))$  differentiable.

# Solving for a symmetric PS-BNE $(s^*)_{j \in I}$

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(a) 
$$u(a_i, v_i) = F((s^*)^{-1}(a_i))^{|I|-1}(v_i - a_i)$$
 differentiable in  $v_i$ .

(b) 
$$U(v_i) = F(v_i)^{|I|-1}(v_i - s^*(v_i))$$
 differentiable.

Use envelope theorem:

$$U'(v_i) = u'_{v_i}(a_i, v_i)|_{a_i = s^*(v_i)} = F((s^*)^{-1}(a_i))^{|I|-1}|_{a_i = s^*(v_i)} = F(v_i)^{|I|-1}.$$
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 (2)

Fundamental theorem of calculus:

$$U(v_i) = U(\underline{v}) + \int_{\underline{v}}^{V_i} U'(v) \, dv = \int_{\underline{v}}^{V_i} U'(v) \, dv = . \tag{3}$$

# Solving for a symmetric PS-BNE $(s^*)_{i \in I}$

Putting it all together:

$$U(v_i) = F(v_i)^{|I|-1}(v_i - s^*(v_i)).$$
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$$U'(v_i) = F(v_i)^{|I|-1}.$$
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$$U(v_i) = \int_{\underline{v}}^{v_i} U'(v) \, \mathrm{d}v. \tag{3}$$

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$$U(v_i) = \int_{v}^{v_i} U'(v) \, \mathrm{d}v. \tag{3}$$

$$\implies \int_{\underline{v}}^{v_i} F(v)^{|I|-1} dv = F(v_i)^{|I|-1} (v_i - s^*(v_i))$$

$$\iff$$
  $s^*(v_i) = v_i - \int_{\underline{v}}^{v_i} \left(\frac{F(v)}{F(v_i)}\right)^{|I|-1} dv$ 

# Solving for a symmetric PS-BNE $(s^*)_{j \in I}$

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Properties of s\*

$$s^*(v'+e)-s^*(v')$$

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Properties of s\*

$$\begin{split} s^*(v'+e) - s^*(v') \\ &= e - \int_{\underline{v}}^{v'+e} \left(\frac{F(v)}{F(v'+e)}\right)^{|I|-1} \, \mathrm{d}v + \int_{\underline{v}}^{v'} \left(\frac{F(v)}{F(v')}\right)^{|I|-1} \, \mathrm{d}v \end{split}$$

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# Solving for a symmetric PS-BNE $(s^*)_{j \in I}$

$$s^*(v_i) = v_i - \int_{v}^{v_i} \left(\frac{F(v)}{F(v_i)}\right)^{|I|-1} dv$$

Properties of s\*

Strictly increasing.

Differentiable (immediate).

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Properties of s\*

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Differentiable (immediate).

Bid less than value  $s^*(v_i) < v_i$  for  $v_i > \underline{v}$ .  $\Longrightarrow U(v_i) \ge 0$ .

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WT check s\* is optimal.

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, then  $0 = u(a_i, v_i) = U(\underline{v}) \le U(v_i)$ .

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, then  $u(a_i, v_i) = v_i - a_i < v_i - s^*(\overline{v})$ .

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If  $v_i > v_i'$ , then  $F(v) \ge F(v_i')$  for any  $v \in [v_i', v_i] \implies U(v_i) - u(a_i, v_i) \ge 0$ .

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# Solving for a symmetric PS-BNE $(s^*)_{j \in I}$

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$$= \int_{v_{i}'}^{v_{i}} \left( F(v)^{|I|-1} - F(v_{i}')^{|I|-1} \right) dv$$

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and  $F(v_i') \ge F(v)$  for any  $v \in [v_i, v_i'] \implies U(v_i) - u(a_i, v_i) \ge 0$ .

## Revenue Equivalence Theorem: Any auction setting such that

- (i) bidders' types are their valuation, drawn independently from compact convex set,
- (ii) the object is allocated to the bidder with the highest valuation,
- (iii) a bidder with the lowest possible valuation ( $\underline{v}$ ) gets 0 in expected payoff in equilibrium

generates the same expected revenue to the auctioneer as the 2PA.

 $V^{k:n}$ : k-th highest valuation out of I bidders.

 $\implies$  Revenue in 2PA:  $V^{2:I}$ .

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 $\implies$  Revenue in 2PA:  $V^{2:l}$ .

Bids in 1PA:

$$s^{*}(v) = v - \int_{\underline{v}}^{v} \left(\frac{F(s)}{F(v)}\right)^{l-1} ds = \frac{1}{F^{1:l-1}(v)} \left[F^{1:l-1}(v)v - \int_{\underline{v}}^{v} F^{1:l-1}(s) ds\right]$$

$$= \frac{1}{F^{1:l-1}(v)} \int_{\underline{v}}^{v} s dF^{1:l-1}(s) \qquad \text{(Integration by parts)}$$

$$= \mathbb{E}[v^{1:l-1}|v^{1:l-1} < v]$$

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$$= \frac{1}{F^{1:l-1}(v)} \int_{\underline{v}}^{v} s dF^{1:l-1}(s) \qquad \text{(Integration by parts)}$$

$$= \mathbb{E}[v^{1:l-1}|v^{1:l-1} < v]$$

⇒ Revenue in 1PA:

$$s^*(V^{1:l}) = \mathbb{E}[V^{1:l-1}|V^{1:l-1} < V^{1:l}]$$

Revenue Equivalence:  $\mathbb{E}[V^{2:l}] = \mathbb{E}[\mathbb{E}[V^{1:l-1}|V^{1:l-1} < V^{1:l}]]$ 

## Overview

- 1. Motivation
- 2. Bayesian Games
- 3. Bayesian Nash Equilibrium
- 4. Auctions
- 5. Purification Theorem
- 6. Higher-Order Beliefs
- 7. More

## **Purification Theorem**

**MSNE hard to justify:** although player is indifferent, they need to randomise in very particular way to make opponents indifferent as well.

**Purification:** Harsanyi (1973) provided a justification for MSNE of a normal-form game  $\Gamma = \langle I, S, u \rangle$  as a limit case of perturbed games.

Suppose true preference is unobserved by opponents (random) and given by

$$\tilde{u}_i(s, \theta_i) := u_i(s) + \varepsilon \theta_i^s$$

where  $\theta_i^s$  are independent across players and drawn from a distribution  $F_i$  with density  $f_i$ .

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where  $\theta_i^s$  are independent across players and drawn from a distribution  $F_i$  with density  $f_i$ .

#### **Theorem**

Fix a finite set of players I and strategy spaces  $S_i$ . For almost all payoff vectors  $u=(u_i)_{i\in I}$  and for all independent and twice-differentiable densities  $f_i$  on  $[-1,1]^{|S|}$ , any mixed strategy Nash equilibrium of the normal-form game  $\Gamma=\langle I,S,u\rangle$  is the limite of a sequence of pure strategy Bayesian Nash equilibria of the Bayesian game with perturbed payoffs  $(\tilde{u}_i)_{i\in I}$ .

## **Purification Theorem**

**MSNE hard to justify:** although player is indifferent, they need to randomise in very particular way to make opponents indifferent as well.

**Purification:** Harsanyi (1973) provided a justification for MSNE of a normal-form game  $\Gamma = \langle I, S, u \rangle$  as a limit case of perturbed games.

Suppose true preference is unobserved by opponents (random) and given by

$$\tilde{u}_i(s, \theta_i) := u_i(s) + \varepsilon \theta_i^s$$

where  $\theta_i^s$  are independent across players and drawn from a distribution  $F_i$  with density  $f_i$ .

#### **Theorem**

Fix a finite set of players I and strategy spaces  $S_i$ . For almost all payoff vectors  $u = (u_i)_{i \in I}$  and for all independent and twice-differentiable densities  $f_i$  on  $[-1,1]^{|S|}$ , any mixed strategy Nash equilibrium of the normal-form game  $\Gamma = \langle I, S, u \rangle$  is the limite of a sequence of pure strategy Bayesian Nash equilibria of the Bayesian game with perturbed payoffs  $(\tilde{u}_i)_{i \in I}$ .

Note the limits of the result: "for almost all payoff vectors"

Coordination Game (bank runs, currency attacks)

		Col Player	
		Invest	Not Invest
Row Player	Invest	$\theta$ , $\theta$	$\theta$ – 1,0
	Not Invest	0.00 - 1	0,0

### Complete Information. NE?

 $\theta$  < 0: Not invest is strictly dominant and (NI,NI) the unique NE.

 $\theta$  > 1: Invest is strictly dominant and (I,I) the unique NE.

 $\theta \in [0, 1]$ : (NI,NI), (I,I), and mixed is NE.

Coordination Game (bank runs, currency attacks)

		Col Player	
		Invest	Not Invest
Row Player	Invest	$\theta$ , $\theta$	$\theta$ – 1,0
	Not Invest	0.00 - 1	0,0

### **Incomplete Information**

Suppose both players observe a signal about the state  $\theta$ .

$$\theta_i := \theta + \epsilon_i$$
,  $\epsilon_i \sim N(0, \sigma^2)$  iid.

$$\theta \mid \theta_i \sim N(\theta_i, \sigma^2)$$
, because  $\theta = \theta_i - \epsilon_i$ 

$$\theta_j \mid \theta_i := \theta \mid \theta_i + \epsilon_j \mid \theta_i = \theta \mid \theta_i + \epsilon_j \sim \textit{N}(\theta_i, 2\sigma^2).$$

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Coordination Game (bank runs, currency attacks)

		Col Player	
		Invest	Not Invest
Row Player	Invest	$\theta,\theta$	$\theta$ – 1,0
	Not Invest	$0,\theta - 1$	0,0

### Incomplete Information

Suppose both players observe a signal about the state  $\theta$ .

$$\theta_i := \theta + \epsilon_i$$
,  $\epsilon_i \sim N(0, \sigma^2)$  iid.

$$\begin{aligned} \theta \mid \theta_i &\sim \textit{N}(\theta_i, \sigma^2), \, \text{because} \, \theta = \theta_i - \epsilon_i \\ \theta_i \mid \theta_i &:= \theta \mid \theta_i + \epsilon_i \mid \theta_i = \theta \mid \theta_i + \epsilon_i \sim \textit{N}(\theta_i, 2\sigma^2). \end{aligned}$$

Implicitly, this is saying that players have uninformative or improper prior on  $\theta$  that is uniform over the real line.

Why improper? because there is no uniform distribution over the real line; it cannot add-up to one if it has a constant pdf.

## **Incomplete Information**

 $\theta \mid \theta_i \sim \textit{N}(\theta_i, \sigma^2) \text{ and } \theta_j \mid \theta_i \sim \textit{N}(\theta_i, 2\sigma^2).$ 

### **Incomplete Information**

 $\theta \mid \theta_i \sim N(\theta_i, \sigma^2)$  and  $\theta_j \mid \theta_i \sim N(\theta_i, 2\sigma^2)$ .

**Claim**:  $s_i(\theta_i) := 1\{\theta_i > 1/2\}$  is an equilibrium.

• Given  $s_j = 1\{\theta_j > 1/2\}$ , player i's payoff to investing conditional on  $\theta_i$  and  $s_j$  is

$$\theta_i - \mathbb{P}\left(\theta_j \leq 1/2 \mid \theta_i\right) = \theta_i - \Phi\left(\frac{1/2 - \theta_i}{\sqrt{2}\sigma}\right)$$

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strictly increasing in  $\theta_i$  and zero when  $\theta_i = 1/2$ .

•  $s_i(\theta_i) := 1\{\theta_i > 1/2\}$  is the unique best response to  $s_j$ .

## **Incomplete Information**

$$\theta \mid \theta_i \sim N(\theta_i, \sigma^2)$$
 and  $\theta_j \mid \theta_i \sim N(\theta_i, 2\sigma^2)$ .

WTS **Proposition**: In any eqm,  $s_i(\theta_i) = 1$  a.e. on  $(1/2, \infty)$  and  $s_i(\theta_i) = 0$  a.e. on  $(-\infty, 1/2)$ .

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$$f(\theta_i, \tilde{\theta}) := \theta_i - \Phi\left(\frac{\tilde{\theta} - \theta_i}{\sqrt{2}\sigma}\right)$$
.

Note: 
$$\mathbb{P}(\theta_j < \tilde{\theta} \mid \theta_i) = \mathbb{P}\left(\frac{\theta_j - \theta_i}{\sqrt{2}\sigma} < \frac{\tilde{\theta} - \theta_i}{\sqrt{2}\sigma} \mid \theta_i\right) = \Phi\left(\frac{\tilde{\theta} - \theta_i}{\sqrt{2}\sigma}\right)$$
.

 $f(\theta_i, \tilde{\theta})$  is continuous in  $(\theta_i, \tilde{\theta})$ , strictly increasing in  $\theta_i$ , and strictly decreasing in  $\tilde{\theta}$ .

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 $\textbf{Claim: } \forall \theta_i > 1, \, \mathbb{E}[u_i(l,s_j(\theta_j),\theta)|\theta_i] = \theta_i - \mathbb{E}[s_j(\theta_j)|\theta_i] > 0.$ 

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- Note that:  $\forall \delta > 0$ ,  $f(\theta_i, \tilde{\theta}) > \delta \ \forall \theta_i \geq 1 + \delta$  and  $\forall \tilde{\theta}$ .
- Then  $\forall \theta_i > 1$ ,  $\mathbb{E}[u_i(l, s_j(\theta_j), \theta) | \theta_i] = \theta_i \mathbb{E}[s_j(\theta_j) | \theta_i] \ge \theta_i = f(\theta_i, -\infty) > 0$  (where  $f(\theta_i, -\infty) := \lim_{\tilde{\theta} \to -\infty} f(\theta_i, \tilde{\theta})$ ).

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$$\theta \mid \theta_i \sim \textit{N}(\theta_i, \sigma^2) \text{ and } \theta_j \mid \theta_i \sim \textit{N}(\theta_i, 2\sigma^2).$$

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For k=1,2,..., define  $\overline{\theta}^{k+1}:=\inf\{\theta_i|f(\theta_i,\overline{\theta}^k)>0\}$ , where  $\overline{\theta}^1=1$ .

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Claim:  $\overline{\theta}^k > \overline{\theta}^{k+1} \forall k$ .

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**Claim:** 
$$\forall \theta_i > 1$$
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Claim:  $\overline{\theta}^k > \overline{\theta}^{k+1} \forall k$ .

- True for k = 0.
- Induction:  $\overline{\theta}^{k+1} < \overline{\theta}^k \implies 0 = f(\overline{\theta}^{k+2}, \overline{\theta}^{k+1}) = f(\overline{\theta}^{k+1}, \overline{\theta}^k) < f(\overline{\theta}^{k+1}, \overline{\theta}^{k+1})$ : f strictly decreasing in 2nd argument and  $\overline{\theta}^{k+1} < \overline{\theta}^k$ .
- $f(\overline{\theta}^{k+2}, \overline{\theta}^{k+1}) < f(\overline{\theta}^{k+1}, \overline{\theta}^{k+1}) \implies \overline{\theta}^{k+2} < \overline{\theta}^{k+1} : f \text{ strictly increasing in 1st argument.}$

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**Claim:** (Induction step) If  $s_j(\theta_j) = 1 \forall \theta_j > \overline{\theta}^k$ , then  $\forall \theta_i > \overline{\theta}^{k+1} \mathbb{E}[u_i(l, s_j(\theta_j), \theta) | \theta_i] > 0$ .

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•  $\forall \theta_i > \overline{\theta}^{k+1}$ ,  $\mathbb{E}[u_i(l, s_j(\theta_j), \theta) | \theta_i] \ge f(\theta_i, \overline{\theta}^k) > 0 = f(\overline{\theta}^{k+1}, \overline{\theta}^k)$  $\therefore f$  is strictly increasing in 1st arg.

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**Claim:** At any BNE s,  $s_i(\theta_i) = 1$  a.e. on  $\theta_i > \overline{\theta}^k$ , for all k and i = 1, 2.

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**Claim:** At any BNE s,  $s_i(\theta_i) = 1$  a.e. on  $\theta_i > \overline{\theta}^k$ , for all k and i = 1, 2.

- True for k = 1,  $\mathbb{E}[u_i(l, s_j(\theta_j), \theta) | \theta_i] \ge \theta_i 1 > 0$  for any  $\theta_i > \overline{\theta}^1 = 1$ , no matter  $s_j$ .
- Then, for any  $s_i': s_i'(\theta_i) \neq 1$  for a positive measure of  $\theta_i > \overline{\theta}^1$  is strictly dominated by  $s_i$  s.t.  $s_i = s_i'$  on  $(-\infty, \overline{\theta}^1]$  and  $s_i = 1$  on  $(\overline{\theta}^1, \infty)$ .
- Iterating the argument, for any  $s_i$ , for any  $s_i'$ :  $s_i'(\theta_i) \neq 1$  for a positive measure of  $\theta_i > \overline{\theta}^k$  is iteratedly strictly dominated by  $s_i$  s.t.  $s_i = s_i'$  on  $(-\infty, \overline{\theta}^k]$  and  $s_i = 1$  on  $(\overline{\theta}^k, \infty)$ .

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 $\theta \mid \theta_i \sim N(\theta_i, \sigma^2)$  and  $\theta_j \mid \theta_i \sim N(\theta_i, 2\sigma^2)$ .

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Claim:  $\forall \theta_i > 1$ ,  $\mathbb{E}[u_i(l, s_j(\theta_j), \theta) | \theta_i] = \theta_i - \mathbb{E}[s_j(\theta_j) | \theta_i] > 0$ .

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Claim:  $\overline{\theta}^k > \overline{\theta}^{k+1} \forall k$ .

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**Claim:** At any BNE s,  $s_i(\theta_i) = 1$  a.e. on  $\theta_i > \overline{\theta}^k$ , for all k and i = 1, 2.

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- Note that:  $\forall \delta > 0$ ,  $f(\theta_i, \tilde{\theta}) < -\delta \forall \theta_i \leq -\delta$  and  $\forall \tilde{\theta}$ .
- Then  $\forall \theta_i < 0$ ,  $\mathbb{E}[u_i(l, s_j(\theta_j), \theta) | \theta_i] = \theta_i \mathbb{E}[s_j(\theta_j) | \theta_i] \le \theta_i = f(\theta_i, \infty) < 0$  (where  $f(\theta_i, \infty) := \lim_{\tilde{\theta} \to \infty} f(\theta_i, \tilde{\theta})$ ).
- Then, as  $f(\overline{\theta}^k, \overline{\theta}^k) > 0 > f(0, \overline{\theta}^k)$  and f is strictly increasing in 1st argument, then  $\overline{\theta}^k > 0 \ \forall k$ .

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- $\{\overline{\theta}^k\}_k$  decreasing sequence, bounded below by  $0 \Longrightarrow$  it converges to some  $\overline{\theta}^\infty \ge 0$ , by monotone convergence theorem.
- $\bullet \ \ 0 = \lim_{k \to \infty} f(\overline{\theta}^{k+1}, \overline{\theta}^k) = f(\overline{\theta}^{\infty}, \overline{\theta}^{\infty}) = \overline{\theta}^{\infty} \Phi\left(\frac{\overline{\theta}^{\infty} \overline{\theta}^{\infty}}{\sqrt{2}\sigma}\right) = \overline{\theta}^{\infty} 1/2.$

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Fully symmetric arguments:

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### Fully symmetric arguments:

**Claim:**  $\forall \theta_i < 0$ ,  $\mathbb{E}[u_i(l, s_i(\theta_i), \theta)|\theta_i] = \theta_i - \mathbb{E}[s_i(\theta_i)|\theta_i] < 0$ .

For  $k = 1, 2, ..., \text{ define } \underline{\theta}^{k+1} := \sup\{\theta_i | f(\theta_i, \underline{\theta}^k) < 0\}, \text{ where } \underline{\theta}^1 = 0.$ 

Claim:  $\theta^k < \theta^{k+1} \forall k$ .

**Claim:** (Induction step) If  $s_j(\theta_j) = 0 \forall \theta_j < \underline{\theta}^k$ , then  $\forall \theta_i < \underline{\theta}^{k+1} \mathbb{E}[u_i(l, s_j(\theta_j), \theta) | \theta_i] < 0$ .

**Claim:** At any BNE s,  $s_i(\theta_i) = 0$  a.e. on  $\theta_i < \underline{\theta}^k$ , for all k and i = 1, 2.

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**Proposition**: In any eqm,  $s_i(\theta_i) = 1$  a.e. on  $(1/2, \infty)$  and  $s_i(\theta_i) = 0$  a.e. on  $(-\infty, 1/2)$ .

NB: proposition holds  $\forall \sigma$ . Taking  $\sigma \downarrow 0$  selects unique NE.

Global game approach to selection of NE.

Why higher-order beliefs?

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Iterating the argument on players' beliefs has higher-order beliefs working in the background to refine what the opponent will do.

Common knowledge of rationality is doing all the heavy-lifting in determining how players behave!

### Overview

- 1. Motivation
- 2. Bayesian Games
- 3. Bayesian Nash Equilibrium
- 4. Auctions
- 5. Purification Theorem
- 6. Higher-Order Beliefs
- 7. More

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Experiments: Winner's curse Charness & Levin (2009 AEJMicro), Overbidding and QRE: Goeree, Holt, & Palfrey (2002 JET); Camerer, Nunnari, & Palfrey (2016 GEB); and Charness, Levin, & Schmeidler (2019 JET).